

## The Hull-White Model

- Hull and White (1987) postulate the following model,

$$\begin{aligned}\frac{dS}{S} &= r dt + \sqrt{V} dW_1, \\ dV &= \mu_v V dt + bV dW_2.\end{aligned}$$

- Above,  $V$  is the instantaneous variance.
- They assume  $\mu_v$  depends on  $V$  and  $t$  (but not  $S$ ).

## The SABR Model

- Hagan, Kumar, Lesniewski, and Woodward (2002) postulate the following model,

$$\begin{aligned}\frac{dS}{S} &= r dt + S^\theta V dW_1, \\ dV &= bV dW_2,\end{aligned}$$

for  $0 \leq \theta \leq 1$ .

- A nice feature of this model is that the implied volatility surface has a compact approximate closed form.

## The Hilliard-Schwartz Model

- Hilliard and Schwartz (1996) postulate the following general model,

$$\begin{aligned}\frac{dS}{S} &= r dt + f(S)V^a dW_1, \\ dV &= \mu(V) dt + bV dW_2,\end{aligned}$$

for some well-behaved function  $f(S)$  and constant  $a$ .

## The Blacher Model

- Blacher (2002) postulates the following model,

$$\begin{aligned}\frac{dS}{S} &= r dt + \sigma \left[ 1 + \alpha(S - S_0) + \beta(S - S_0)^2 \right] dW_1, \\ d\sigma &= \kappa(\theta - \sigma) dt + \epsilon\sigma dW_2.\end{aligned}$$

- So the volatility  $\sigma$  follows a mean-reverting process to level  $\theta$ .

## Heston's Stochastic-Volatility Model

- Heston (1993) assumes the stock price follows

$$\frac{dS}{S} = (\mu - q) dt + \sqrt{V} dW_1, \quad (64)$$

$$dV = \kappa(\theta - V) dt + \sigma\sqrt{V} dW_2. \quad (65)$$

- $V$  is the instantaneous variance, which follows a square-root process.
  - $dW_1$  and  $dW_2$  have correlation  $\rho$ .
  - The riskless rate  $r$  is constant.
- It may be the most popular continuous-time stochastic-volatility model.<sup>a</sup>

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<sup>a</sup>Christoffersen, Heston, and Jacobs (2009).

## Heston's Stochastic-Volatility Model (continued)

- Heston assumes the market price of risk is  $b_2\sqrt{V}$ .
- So  $\mu = r + b_2V$ .
- Define

$$\begin{aligned}dW_1^* &= dW_1 + b_2\sqrt{V} dt, \\dW_2^* &= dW_2 + \rho b_2\sqrt{V} dt, \\ \kappa^* &= \kappa + \rho b_2\sigma, \\ \theta^* &= \frac{\theta\kappa}{\kappa + \rho b_2\sigma}.\end{aligned}$$

- $dW_1^*$  and  $dW_2^*$  have correlation  $\rho$ .

## Heston's Stochastic-Volatility Model (continued)

- Under the risk-neutral probability measure  $Q$ , both  $W_1^*$  and  $W_2^*$  are Wiener processes.
- Heston's model becomes, under probability measure  $Q$ ,

$$\begin{aligned}\frac{dS}{S} &= (r - q) dt + \sqrt{V} dW_1^*, \\ dV &= \kappa^*(\theta^* - V) dt + \sigma\sqrt{V} dW_2^*.\end{aligned}$$

## Heston's Stochastic-Volatility Model (continued)

- Define

$$\begin{aligned}\phi(u, \tau) = & \exp \left\{ \imath u (\ln S + (r - q) \tau) \right. \\ & + \theta^* \kappa^* \sigma^{-2} \left[ (\kappa^* - \rho \sigma \imath u - d) \tau - 2 \ln \frac{1 - g e^{-d\tau}}{1 - g} \right] \\ & \left. + \frac{v \sigma^{-2} (\kappa^* - \rho \sigma \imath u - d) (1 - e^{-d\tau})}{1 - g e^{-d\tau}} \right\},\end{aligned}$$

$$d = \sqrt{(\rho \sigma \imath u - \kappa^*)^2 - \sigma^2 (-\imath u - u^2)},$$

$$g = (\kappa^* - \rho \sigma \imath u - d) / (\kappa^* - \rho \sigma \imath u + d).$$

## Heston's Stochastic-Volatility Model (concluded)

The formulas are<sup>a</sup>

$$\begin{aligned}
 C &= S \left[ \frac{1}{2} + \frac{1}{\pi} \int_0^\infty \operatorname{Re} \left( \frac{X^{-\imath u} \phi(u - \imath, \tau)}{\imath u S e^{r\tau}} \right) du \right] \\
 &\quad - X e^{-r\tau} \left[ \frac{1}{2} + \frac{1}{\pi} \int_0^\infty \operatorname{Re} \left( \frac{X^{-\imath u} \phi(u, \tau)}{\imath u} \right) du \right], \\
 P &= X e^{-r\tau} \left[ \frac{1}{2} - \frac{1}{\pi} \int_0^\infty \operatorname{Re} \left( \frac{X^{-\imath u} \phi(u, \tau)}{\imath u} \right) du \right], \\
 &\quad - S \left[ \frac{1}{2} - \frac{1}{\pi} \int_0^\infty \operatorname{Re} \left( \frac{X^{-\imath u} \phi(u - \imath, \tau)}{\imath u S e^{r\tau}} \right) du \right],
 \end{aligned}$$

where  $\imath = \sqrt{-1}$  and  $\operatorname{Re}(x)$  denotes the real part of the complex number  $x$ .

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<sup>a</sup>Contributed by Mr. Chen, Chun-Ying (D95723006) on August 17, 2008 and Mr. Liou, Yan-Fu (R92723060) on August 26, 2008.

## Stochastic-Volatility Models and Further Extensions<sup>a</sup>

- How to explain the October 1987 crash?
- Stochastic-volatility models require an implausibly high-volatility level prior to *and* after the crash.
- Merton (1976) proposed jump models.
- Discontinuous jump models *in the asset price* can alleviate the problem somewhat.

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<sup>a</sup>Eraker (2004).

## Stochastic-Volatility Models and Further Extensions (continued)

- But if the jump intensity is a constant, it cannot explain the tendency of large movements to cluster over time.
- This assumption also has no impacts on option prices.
- Jump-diffusion models combine both.
  - E.g., add a jump process to Eq. (64) on p. 603.
  - Closed-form formulas exist for GARCH-jump option pricing models.<sup>a</sup>

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<sup>a</sup>Liou (R92723060) (2005).

## Stochastic-Volatility Models and Further Extensions (concluded)

- But they still do not adequately describe the systematic variations in option prices.<sup>a</sup>
- Jumps *in volatility* are alternatives.<sup>b</sup>
  - E.g., add correlated jump processes to Eqs. (64) and Eq. (65) on p. 603.
- Such models allow high level of volatility caused by a jump to volatility.<sup>c</sup>

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<sup>a</sup>Bates (2000) and Pan (2002).

<sup>b</sup>Duffie, Pan, and Singleton (2000).

<sup>c</sup>Eraker, Johnnes, and Polson (2000).

## Complexities of Stochastic-Volatility Models

- A few stochastic-volatility models suffer from subexponential ( $c^{\sqrt{n}}$ ) tree size.
- Examples include the Hull-White (1987), Hilliard-Schwartz (1996), and SABR (2002) models.<sup>a</sup>
- Future research may extend this negative result to more stochastic-volatility models.
  - We suspect many GARCH option pricing models entertain similar problems.<sup>b</sup>

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<sup>a</sup>Chiu (R98723059) (2012).

<sup>b</sup>Chen (R95723051) (2008); Chen (R95723051), Lyuu, and Wen (D94922003) (2011).

# *Hedging*

When Professors Scholes and Merton and I  
invested in warrants,  
Professor Merton lost the most money.  
And I lost the least.  
— Fischer Black (1938–1995)

## Delta Hedge

- The delta (hedge ratio) of a derivative  $f$  is defined as

$$\Delta \equiv \frac{\partial f}{\partial S}.$$

- Thus

$$\Delta f \approx \Delta \times \Delta S$$

for relatively small changes in the stock price,  $\Delta S$ .

- A delta-neutral portfolio is hedged as it is immunized against small changes in the stock price.
- A trading strategy that dynamically maintains a delta-neutral portfolio is called delta hedge.

## Delta Hedge (concluded)

- Delta changes with the stock price.
- A delta hedge needs to be rebalanced periodically in order to maintain delta neutrality.
- In the limit where the portfolio is adjusted continuously, “perfect” hedge is achieved and the strategy becomes self-financing.

## Implementing Delta Hedge

- We want to hedge  $N$  *short* derivatives.
- Assume the stock pays no dividends.
- The delta-neutral portfolio maintains  $N \times \Delta$  shares of stock plus  $B$  borrowed dollars such that

$$-N \times f + N \times \Delta \times S - B = 0.$$

- At next rebalancing point when the delta is  $\Delta'$ , buy  $N \times (\Delta' - \Delta)$  shares to maintain  $N \times \Delta'$  shares with a total borrowing of  $B' = N \times \Delta' \times S' - N \times f'$ .
- Delta hedge is the discrete-time analog of the continuous-time limit and will rarely be self-financing.

## Example

- A hedger is *short* 10,000 European calls.
- $S = 50$ ,  $\sigma = 30\%$ , and  $r = 6\%$ .
- This call's expiration is four weeks away, its strike price is \$50, and each call has a current value of  $f = 1.76791$ .
- As an option covers 100 shares of stock,  $N = 1,000,000$ .
- The trader adjusts the portfolio weekly.
- The calls are replicated well if the cumulative cost of trading *stock* is close to the call premium's FV.<sup>a</sup>

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<sup>a</sup>This example takes the replication viewpoint.

## Example (continued)

- As  $\Delta = 0.538560$

$$N \times \Delta = 538,560$$

shares are purchased for a total cost of

$$538,560 \times 50 = 26,928,000$$

dollars to make the portfolio delta-neutral.

- The trader finances the purchase by borrowing

$$B = N \times \Delta \times S - N \times f = 25,160,090$$

dollars net.<sup>a</sup>

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<sup>a</sup>This takes the hedging viewpoint — an alternative. See an exercise in the text.

### Example (continued)

- At 3 weeks to expiration, the stock price rises to \$51.
- The new call value is  $f' = 2.10580$ .
- So the portfolio is worth

$$-N \times f' + 538,560 \times 51 - Be^{0.06/52} = 171,622$$

before rebalancing.

## Example (continued)

- A delta hedge does not replicate the calls perfectly; it is not self-financing as \$171,622 can be withdrawn.
- The magnitude of the tracking error—the variation in the net portfolio value—can be mitigated if adjustments are made more frequently.
- In fact, the tracking error *over one rebalancing act* is positive about 68% of the time, but its expected value is essentially zero.<sup>a</sup>
- The tracking error at maturity is proportional to vega.<sup>b</sup>

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<sup>a</sup>Boyle and Emanuel (1980).

<sup>b</sup>Kamal and Derman (1999).

## Example (continued)

- In practice tracking errors will cease to decrease beyond a certain rebalancing frequency.

- With a higher delta  $\Delta' = 0.640355$ , the trader buys

$$N \times (\Delta' - \Delta) = 101,795$$

shares for \$5,191,545.

- The number of shares is increased to  $N \times \Delta' = 640,355$ .

### Example (continued)

- The cumulative cost is

$$26,928,000 \times e^{0.06/52} + 5,191,545 = 32,150,634.$$

- The portfolio is again delta-neutral.

$\tau$	$S$	Option value $f$	Delta $\Delta$	Change in delta	No. shares bought $N \times (5)$	Cost of shares $(1) \times (6)$	Cumulative cost $FV(8') + (7)$
	(1)	(2)	(3)	(5)	(6)	(7)	(8)
4	50	1.7679	0.53856	—	538,560	26,928,000	26,928,000
3	51	2.1058	0.64036	0.10180	101,795	5,191,545	32,150,634
2	53	3.3509	0.85578	0.21542	215,425	11,417,525	43,605,277
1	52	2.2427	0.83983	-0.01595	-15,955	-829,660	42,825,960
0	54	4.0000	1.00000	0.16017	160,175	8,649,450	51,524,853

The total number of shares is 1,000,000 at expiration (trading takes place at expiration, too).

## Example (concluded)

- At expiration, the trader has 1,000,000 shares.
- They are exercised against by the in-the-money calls for \$50,000,000.
- The trader is left with an obligation of

$$51,524,853 - 50,000,000 = 1,524,853,$$

which represents the replication cost.

- Compared with the FV of the call premium,

$$1,767,910 \times e^{0.06 \times 4/52} = 1,776,088,$$

the net gain is  $1,776,088 - 1,524,853 = 251,235$ .

## Tracking Error Revisited

- Define the dollar gamma as  $S^2\Gamma$ .
- The change in value of a delta-hedged *long* option position after a duration of  $\Delta t$  is proportional to the dollar gamma.
- It is about

$$(1/2)S^2\Gamma[(\Delta S/S)^2 - \sigma^2\Delta t].$$

–  $(\Delta S/S)^2$  is called the daily realized variance.

## Tracking Error Revisited (continued)

- Let the rebalancing times be  $t_1, t_2, \dots, t_n$ .
- Let  $\Delta S_i = S_{i+1} - S_i$ .
- The total tracking error at expiration is about

$$\sum_{i=0}^{n-1} e^{r(T-t_i)} \frac{S_i^2 \Gamma_i}{2} \left[ \left( \frac{\Delta S_i}{S_i} \right)^2 - \sigma^2 \Delta t \right],$$

- The tracking error is path dependent.

## Tracking Error Revisited (concluded)<sup>a</sup>

- The tracking error  $\epsilon_n$  over  $n$  rebalancing acts (such as 251,235 on p. 624) has about the same probability of being positive as being negative.
- Subject to certain regularity conditions, the root-mean-square tracking error  $\sqrt{E[\epsilon_n^2]}$  is  $O(1/\sqrt{n})$ .<sup>b</sup>
- The root-mean-square tracking error increases with  $\sigma$  at first and then decreases.

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<sup>a</sup>Bertsimas, Kogan, and Lo (2000).

<sup>b</sup>See also Grannan and Swindle (1996).

## Delta-Gamma Hedge

- Delta hedge is based on the first-order approximation to changes in the derivative price,  $\Delta f$ , due to changes in the stock price,  $\Delta S$ .
- When  $\Delta S$  is not small, the second-order term, gamma  $\Gamma \equiv \partial^2 f / \partial S^2$ , helps (theoretically).<sup>a</sup>
- A delta-gamma hedge is a delta hedge that maintains zero portfolio gamma, or gamma neutrality.
- To meet this extra condition, one more security needs to be brought in.

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<sup>a</sup>See the numerical example on pp. 231–232 of the text.

## Delta-Gamma Hedge (concluded)

- Suppose we want to hedge short calls as before.
- A hedging call  $f_2$  is brought in.
- To set up a delta-gamma hedge, we solve

$$\begin{aligned} -N \times f + n_1 \times S + n_2 \times f_2 - B &= 0 && \text{(self-financing),} \\ -N \times \Delta + n_1 + n_2 \times \Delta_2 - 0 &= 0 && \text{(delta neutrality),} \\ -N \times \Gamma + 0 + n_2 \times \Gamma_2 - 0 &= 0 && \text{(gamma neutrality),} \end{aligned}$$

for  $n_1$ ,  $n_2$ , and  $B$ .

- The gammas of the stock and bond are 0.

## Other Hedges

- If volatility changes, delta-gamma hedge may not work well.
- An enhancement is the delta-gamma-vega hedge, which also maintains vega zero portfolio vega.
- To accomplish this, one more security has to be brought into the process.
- In practice, delta-vega hedge, which may not maintain gamma neutrality, performs better than delta hedge.

# *Trees*

I love a tree more than a man.  
— Ludwig van Beethoven (1770–1827)

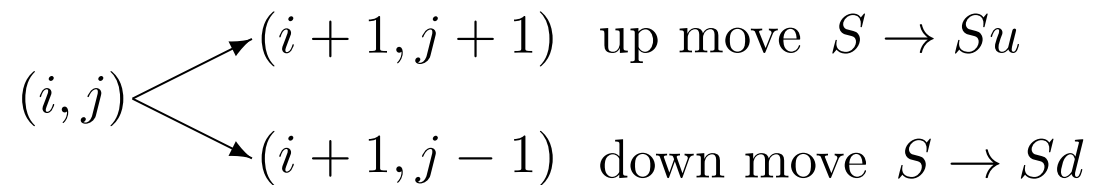
And though the holes were rather small,  
they had to count them all.  
— The Beatles, *A Day in the Life* (1967)

## The Combinatorial Method

- The combinatorial method can often cut the running time by an order of magnitude.
- The basic paradigm is to count the number of admissible paths that lead from the root to any terminal node.
- We first used this method in the linear-time algorithm for standard European option pricing on p. 261.
- We will now apply it to price barrier options.

## The Reflection Principle<sup>a</sup>

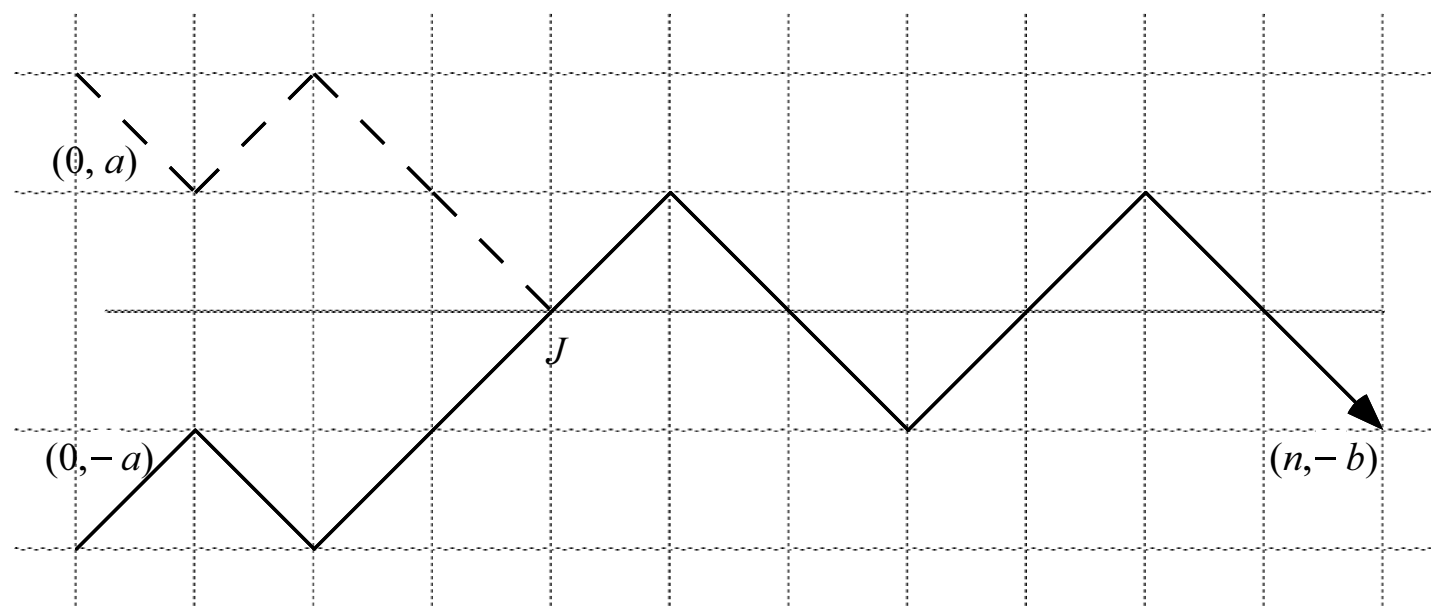
- Imagine a particle at position  $(0, -\mathbf{a})$  on the integral lattice that is to reach  $(n, -\mathbf{b})$ .
- Without loss of generality, assume  $\mathbf{a} > 0$  and  $\mathbf{b} \geq 0$ .
- This particle's movement:



- How many paths touch the  $x$  axis?

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<sup>a</sup>André (1887).



## The Reflection Principle (continued)

- For a path from  $(0, -a)$  to  $(n, -b)$  that touches the  $x$  axis, let  $J$  denote the first point this happens.
- Reflect the portion of the path from  $(0, -a)$  to  $J$ .
- A path from  $(0, a)$  to  $(n, -b)$  is constructed.
- It also hits the  $x$  axis at  $J$  for the first time.
- The one-to-one mapping shows the number of paths from  $(0, -a)$  to  $(n, -b)$  that touch the  $x$  axis equals the number of paths from  $(0, a)$  to  $(n, -b)$ .

## The Reflection Principle (concluded)

- A path of this kind has  $(n + \mathbf{b} + \mathbf{a})/2$  down moves and  $(n - \mathbf{b} - \mathbf{a})/2$  up moves.<sup>a</sup>
- Hence there are

$$\binom{n}{\frac{n+\mathbf{a}+\mathbf{b}}{2}} = \binom{n}{\frac{n-\mathbf{a}-\mathbf{b}}{2}} \quad (66)$$

such paths for even  $n + \mathbf{a} + \mathbf{b}$ .

– Convention:  $\binom{n}{k} = 0$  for  $k < 0$  or  $k > n$ .

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<sup>a</sup>Verify it!

## Pricing Barrier Options (Lyu, 1998)

- Focus on the down-and-in call with barrier  $H < X$ .
- Assume  $H < S$  without loss of generality.
- Define

$$a \equiv \left\lceil \frac{\ln(X/(Sd^n))}{\ln(u/d)} \right\rceil = \left\lceil \frac{\ln(X/S)}{2\sigma\sqrt{\Delta t}} + \frac{n}{2} \right\rceil,$$
$$h \equiv \left\lfloor \frac{\ln(H/(Sd^n))}{\ln(u/d)} \right\rfloor = \left\lfloor \frac{\ln(H/S)}{2\sigma\sqrt{\Delta t}} + \frac{n}{2} \right\rfloor.$$

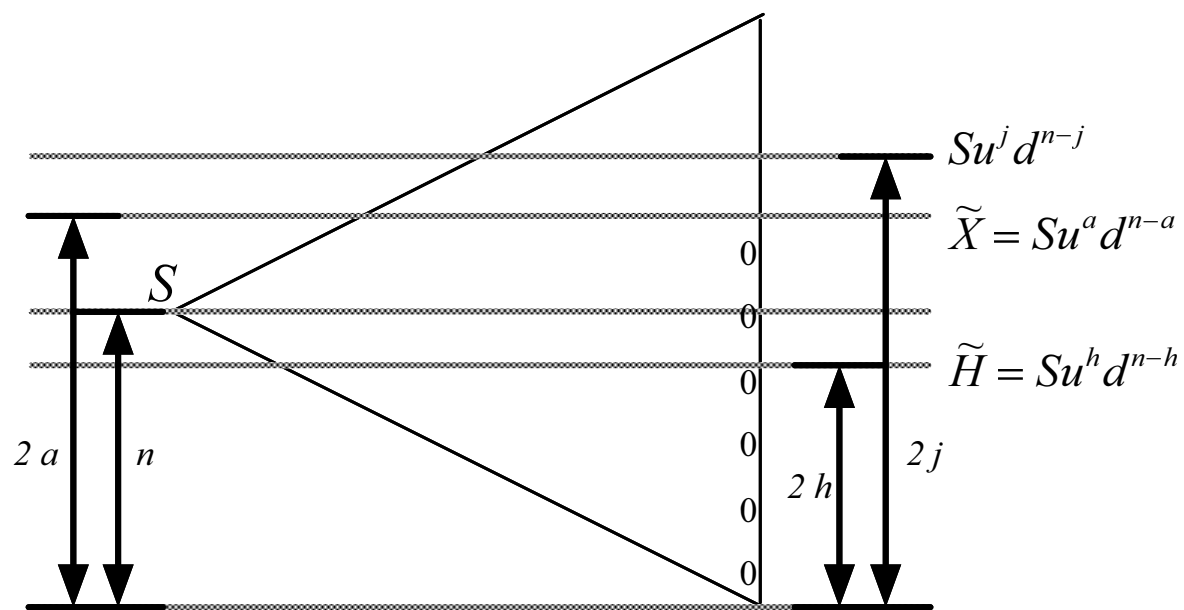
- $a$  is such that  $\tilde{X} \equiv Su^a d^{n-a}$  is the terminal price that is closest to  $X$  from above.
- $h$  is such that  $\tilde{H} \equiv Su^h d^{n-h}$  is the *terminal* price that is closest to  $H$  from below.

## Pricing Barrier Options (continued)

- The true barrier is replaced by the effective barrier  $\tilde{H}$  in the binomial model.
- A process with  $n$  moves hence ends up in the money if and only if the number of up moves is at least  $a$ .
- The price  $Su^k d^{n-k}$  is at a distance of  $2k$  from the lowest possible price  $Sd^n$  on the binomial tree.

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$$Su^k d^{n-k} = Sd^{-k} d^{n-k} = Sd^{n-2k}. \quad (67)$$



## Pricing Barrier Options (continued)

- The number of paths from  $S$  to the terminal price  $Su^j d^{n-j}$  is  $\binom{n}{j}$ , each with probability  $p^j(1-p)^{n-j}$ .
- The reflection principle (p. 640) can be applied with

$$a = n - 2h,$$

$$b = 2j - 2h,$$

in Eq. (66) on p. 637 by treating the  $\tilde{H}$  line as the  $x$  axis.

- Therefore,

$$\binom{n}{\frac{n+(n-2h)+(2j-2h)}{2}} = \binom{n}{n-2h+j}$$

paths hit  $\tilde{H}$  in the process for  $h \leq n/2$ .

## Pricing Barrier Options (concluded)

- The terminal price  $Su^j d^{n-j}$  is reached by a path that hits the effective barrier with probability

$$\binom{n}{n-2h+j} p^j (1-p)^{n-j}, \quad j \leq 2h.$$

- The option value equals

$$\frac{\sum_{j=a}^{2h} \binom{n}{n-2h+j} p^j (1-p)^{n-j} (Su^j d^{n-j} - X)}{R^n}. \quad (68)$$

–  $R \equiv e^{r\tau/n}$  is the riskless return per period.

- It implies a linear-time algorithm.<sup>a</sup>

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<sup>a</sup>Lyuu (1998).

## Convergence of BOPM

- Equation (68) results in the sawtooth-like convergence shown on p. 375.
- The reasons are not hard to see.
- The true barrier most likely does not equal the effective barrier.
- The same holds between the strike price and the effective strike price.
- The issue of the strike price is less critical.
- But the issue of the barrier is not negligible.

## Convergence of BOPM (continued)

- Convergence is actually good if we limit  $n$  to certain values—191, for example.
- These values make the true barrier coincide with or just above one of the stock price levels, that is,

$$H \approx Sd^j = Se^{-j\sigma\sqrt{\tau/n}}$$

for some integer  $j$ .

- The preferred  $n$ 's are thus

$$n = \left\lceil \frac{\tau}{(\ln(S/H)/(j\sigma))^2} \right\rceil, \quad j = 1, 2, 3, \dots$$

## Convergence of BOPM (continued)

- There is only one minor technicality left.
- We picked the effective barrier to be one of the  $n + 1$  possible *terminal* stock prices.
- However, the effective barrier above,  $Sd^j$ , corresponds to a terminal stock price only when  $n - j$  is even.<sup>a</sup>
- To close this gap, we decrement  $n$  by one, if necessary, to make  $n - j$  an even number.

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<sup>a</sup>This is because  $j = n - 2k$  for some  $k$  by Eq. (67) on p. 639. Of course we could have adopted the form  $Sd^j$  ( $-n \leq j \leq n$ ) for the effective barrier. It makes a good exercise.

## Convergence of BOPM (concluded)

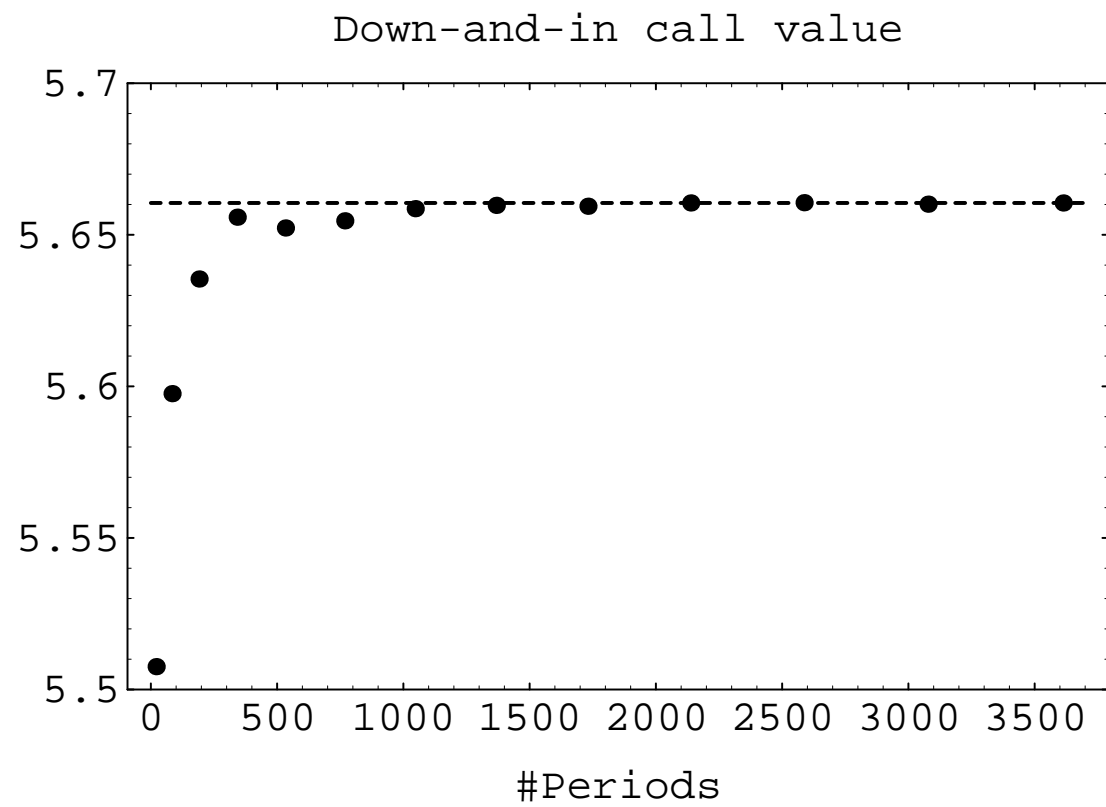
- The preferred  $n$ 's are now

$$n = \begin{cases} \ell & \text{if } \ell - j \text{ is even} \\ \ell - 1 & \text{otherwise} \end{cases},$$

$j = 1, 2, 3, \dots$ , where

$$\ell \equiv \left\lfloor \frac{\tau}{(\ln(S/H)/(j\sigma))^2} \right\rfloor.$$

- Evaluate pricing formula (68) on p. 642 only with the  $n$ 's above.



## Practical Implications

- Now that barrier options can be efficiently priced, we can afford to pick very large  $n$ 's (p. 649).
- This has profound consequences (to explain later on pp. 661ff).

$n$	Combinatorial method	
	Value	Time (milliseconds)
21	5.507548	0.30
84	5.597597	0.90
191	5.635415	2.00
342	5.655812	3.60
533	5.652253	5.60
768	5.654609	8.00
1047	5.658622	11.10
1368	5.659711	15.00
1731	5.659416	19.40
2138	5.660511	24.70
2587	5.660592	30.20
3078	5.660099	36.70
3613	5.660498	43.70
4190	5.660388	44.10
4809	5.659955	51.60
5472	5.660122	68.70
6177	5.659981	76.70
6926	5.660263	86.90
7717	5.660272	97.20

## Practical Implications (concluded)

- Pricing is prohibitively time consuming when  $S \approx H$  because

$$n \sim 1/\ln^2(S/H).$$

- This is called the barrier-too-close problem.
- This observation is indeed true of standard quadratic-time binomial tree algorithms.
- But it no longer applies to linear-time algorithms (see p. 651).
- This binomial model is  $O(1/\sqrt{n})$  convergent in general but  $O(1/n)$  convergent when the barrier is matched.<sup>a</sup>

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<sup>a</sup>Lin (R95221010) (2008) and Lin (R95221010) and Palmer (2010).

Barrier at 95.0			Barrier at 99.5			Barrier at 99.9		
$n$	Value	Time	$n$	Value	Time	$n$	Value	Time
	.							
	.		795	7.47761	8	19979	8.11304	253
2743	2.56095	31.1	3184	7.47626	38	79920	8.11297	1013
3040	2.56065	35.5	7163	7.47682	88	179819	8.11300	2200
3351	2.56098	40.1	12736	7.47661	166	319680	8.11299	4100
3678	2.56055	43.8	19899	7.47676	253	499499	8.11299	6300
4021	2.56152	48.1	28656	7.47667	368	719280	8.11299	8500
True	2.5615			7.4767			8.1130	

(All times in milliseconds.)

## Trinomial Tree

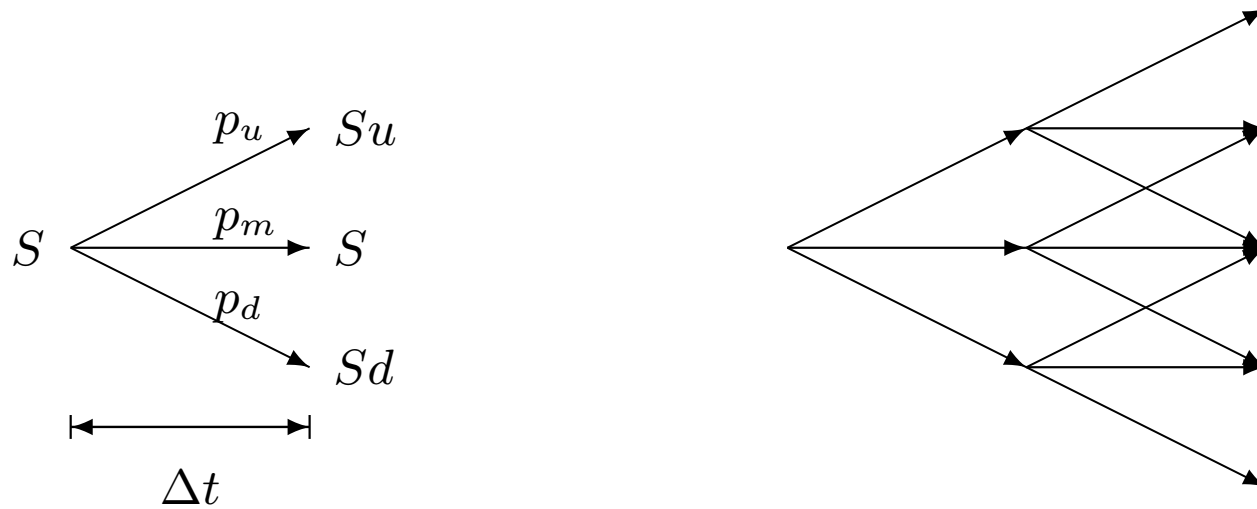
- Set up a trinomial approximation to the geometric Brownian motion<sup>a</sup>

$$\frac{dS}{S} = r dt + \sigma dW.$$

- The three stock prices at time  $\Delta t$  are  $S$ ,  $Su$ , and  $Sd$ , where  $ud = 1$ .
- Let the mean and variance of the stock price be  $SM$  and  $S^2V$ , respectively.

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<sup>a</sup>Boyle (1988).



## Trinomial Tree (continued)

- By Eqs. (21) on p. 161,

$$\begin{aligned}M &\equiv e^{r\Delta t}, \\V &\equiv M^2(e^{\sigma^2\Delta t} - 1).\end{aligned}$$

- Impose the matching of mean and that of variance:

$$\begin{aligned}1 &= p_u + p_m + p_d, \\SM &= (p_u u + p_m + (p_d/u)) S, \\S^2V &= p_u(Su - SM)^2 + p_m(S - SM)^2 + p_d(Sd - SM)^2.\end{aligned}$$

## Trinomial Tree (concluded)

- Use linear algebra to verify that

$$p_u = \frac{u(V + M^2 - M) - (M - 1)}{(u - 1)(u^2 - 1)},$$
$$p_d = \frac{u^2(V + M^2 - M) - u^3(M - 1)}{(u - 1)(u^2 - 1)}.$$

- In practice, we must also make sure the probabilities lie between 0 and 1.
- Countless variations.

## A Trinomial Tree

- Use  $u = e^{\lambda\sigma\sqrt{\Delta t}}$ , where  $\lambda \geq 1$  is a tunable parameter.
- Then

$$\begin{aligned}p_u &\rightarrow \frac{1}{2\lambda^2} + \frac{(r + \sigma^2) \sqrt{\Delta t}}{2\lambda\sigma}, \\p_d &\rightarrow \frac{1}{2\lambda^2} - \frac{(r - 2\sigma^2) \sqrt{\Delta t}}{2\lambda\sigma}.\end{aligned}$$

- A nice choice for  $\lambda$  is  $\sqrt{\pi/2}$ .<sup>a</sup>

---

<sup>a</sup>Omberg (1988).

## Barrier Options Revisited

- BOPM introduces a specification error by replacing the barrier with a nonidentical effective barrier.
- The trinomial model solves the problem by adjusting  $\lambda$  so that the barrier is hit exactly.<sup>a</sup>
- When

$$Se^{-h\lambda\sigma\sqrt{\Delta t}} = H,$$

it takes  $h$  down moves to go from  $S$  to  $H$ , if  $h$  is an integer.

- Then

$$h = \frac{\ln(S/H)}{\lambda\sigma\sqrt{\Delta t}}.$$

---

<sup>a</sup>Ritchken (1995).

## Barrier Options Revisited (continued)

- This is easy to achieve by adjusting  $\lambda$ .
- Typically, we find the smallest  $\lambda \geq 1$  such that  $h$  is an integer.<sup>a</sup>
- That is, we find the largest integer  $j \geq 1$  that satisfies  $\frac{\ln(S/H)}{j\sigma\sqrt{\Delta t}} \geq 1$  and then let

$$\lambda = \frac{\ln(S/H)}{j\sigma\sqrt{\Delta t}}.$$

- Such a  $\lambda$  may not exist for very small  $n$ 's.
- This is not hard to check.

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<sup>a</sup>Why must  $\lambda \geq 1$ ?

## Barrier Options Revisited (concluded)

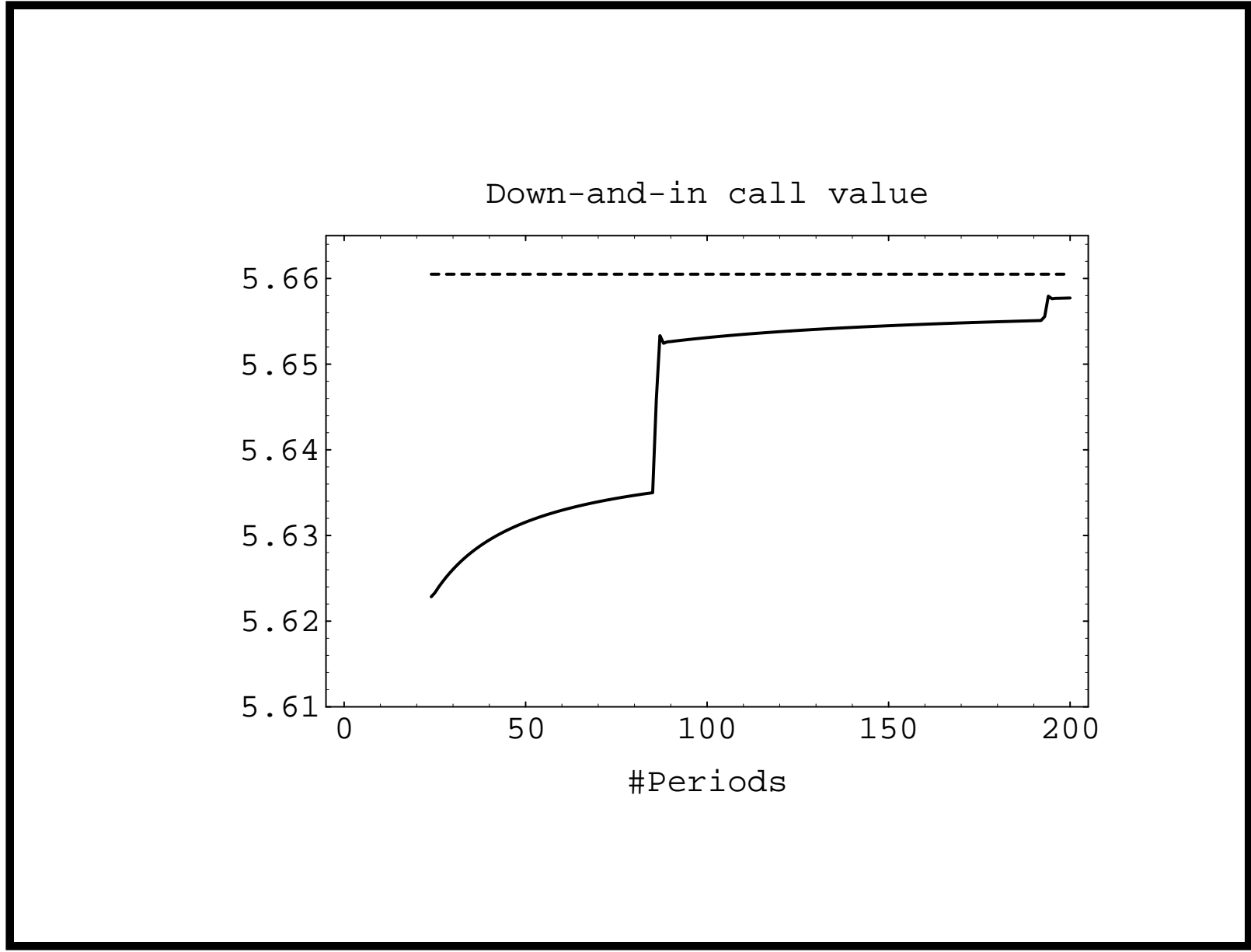
- This done, one of the layers of the trinomial tree coincides with the barrier.
- The following probabilities may be used,

$$p_u = \frac{1}{2\lambda^2} + \frac{\mu' \sqrt{\Delta t}}{2\lambda\sigma},$$

$$p_m = 1 - \frac{1}{\lambda^2},$$

$$p_d = \frac{1}{2\lambda^2} - \frac{\mu' \sqrt{\Delta t}}{2\lambda\sigma}.$$

$$- \mu' \equiv r - \sigma^2/2.$$



## Algorithms Comparison<sup>a</sup>

- So which algorithm is better, binomial or trinomial?
- Algorithms are often compared based on the  $n$  value at which they converge.
  - The one with the smallest  $n$  wins.
- So giraffes are faster than cheetahs because they take fewer strides to travel the same distance!
- Performance must be based on actual running times, not  $n$ .<sup>b</sup>

---

<sup>a</sup>Lyu (1998).

<sup>b</sup>Patterson and Hennessy (1994).

## Algorithms Comparison (continued)

- Pages 375 and 660 seem to show the trinomial model converges at a smaller  $n$  than BOPM.
- It is in this sense when people say trinomial models converge faster than binomial ones.
- But does it make the trinomial model better then?

## Algorithms Comparison (concluded)

- The linear-time binomial tree algorithm actually performs better than the trinomial one.
- See the next page, expanded from p. 649.
- The barrier-too-close problem is also too hard for a quadratic-time trinomial tree algorithm.<sup>a</sup>
- In fact, the trinomial model also has a linear-time algorithm!<sup>b</sup>

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<sup>a</sup>Lyu (1998).

<sup>b</sup>Chen (R94922003) (2007).

$n$	Combinatorial method		Trinomial tree algorithm	
	Value	Time	Value	Time
21	5.507548	0.30		
84	5.597597	0.90	5.634936	35.0
191	5.635415	2.00	5.655082	185.0
342	5.655812	3.60	5.658590	590.0
533	5.652253	5.60	5.659692	1440.0
768	5.654609	8.00	5.660137	3080.0
1047	5.658622	11.10	5.660338	5700.0
1368	5.659711	15.00	5.660432	9500.0
1731	5.659416	19.40	5.660474	15400.0
2138	5.660511	24.70	5.660491	23400.0
2587	5.660592	30.20	5.660493	34800.0
3078	5.660099	36.70	5.660488	48800.0
3613	5.660498	43.70	5.660478	67500.0
4190	5.660388	44.10	5.660466	92000.0
4809	5.659955	51.60	5.660454	130000.0
5472	5.660122	68.70		
6177	5.659981	76.70		

(All times in milliseconds.)

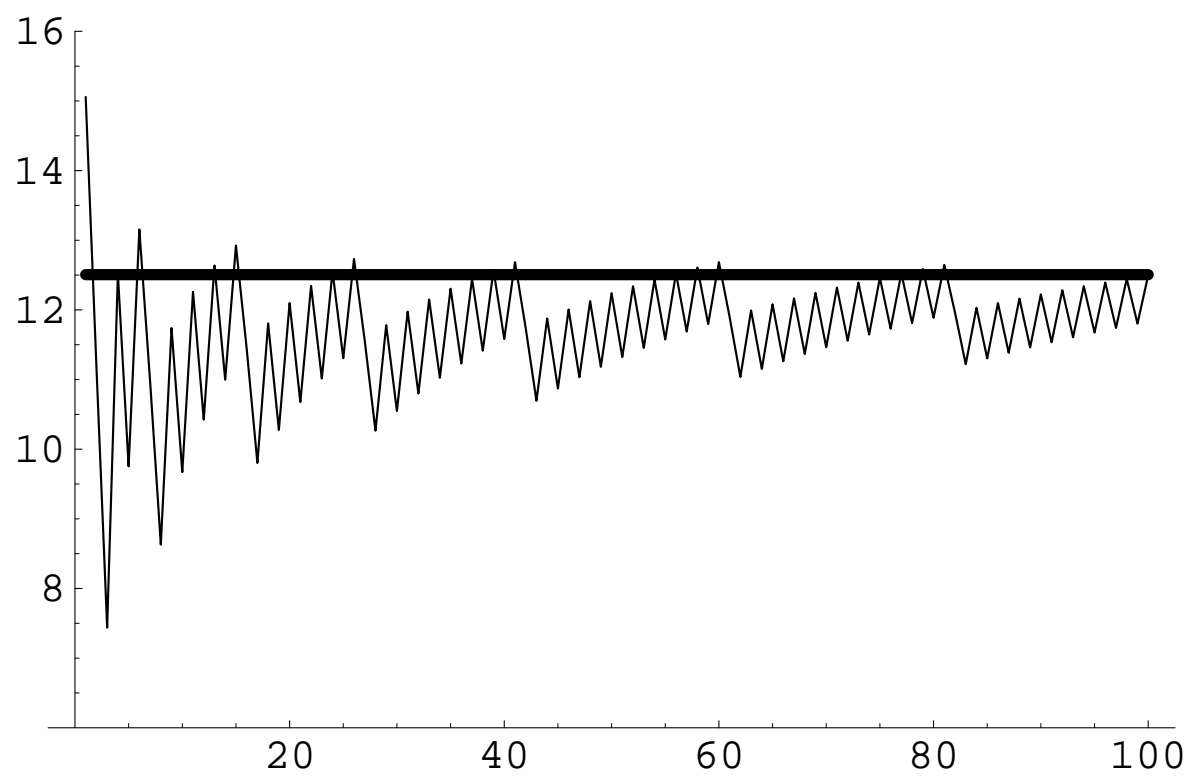
## Double-Barrier Options

- Double-barrier options are barrier options with two barriers  $L < H$ .
- Assume  $L < S < H$ .
- The binomial model produces oscillating option values (see plot on next page).<sup>a</sup>
- The combinatorial method yields a linear-time algorithm (see text).
- This binomial model is  $O(1/\sqrt{n})$  convergent in general.<sup>b</sup>

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<sup>a</sup>Chao (R86526053) (1999); Dai (R86526008, D8852600) and Lyuu (2005).

<sup>b</sup>Gobet (1999).



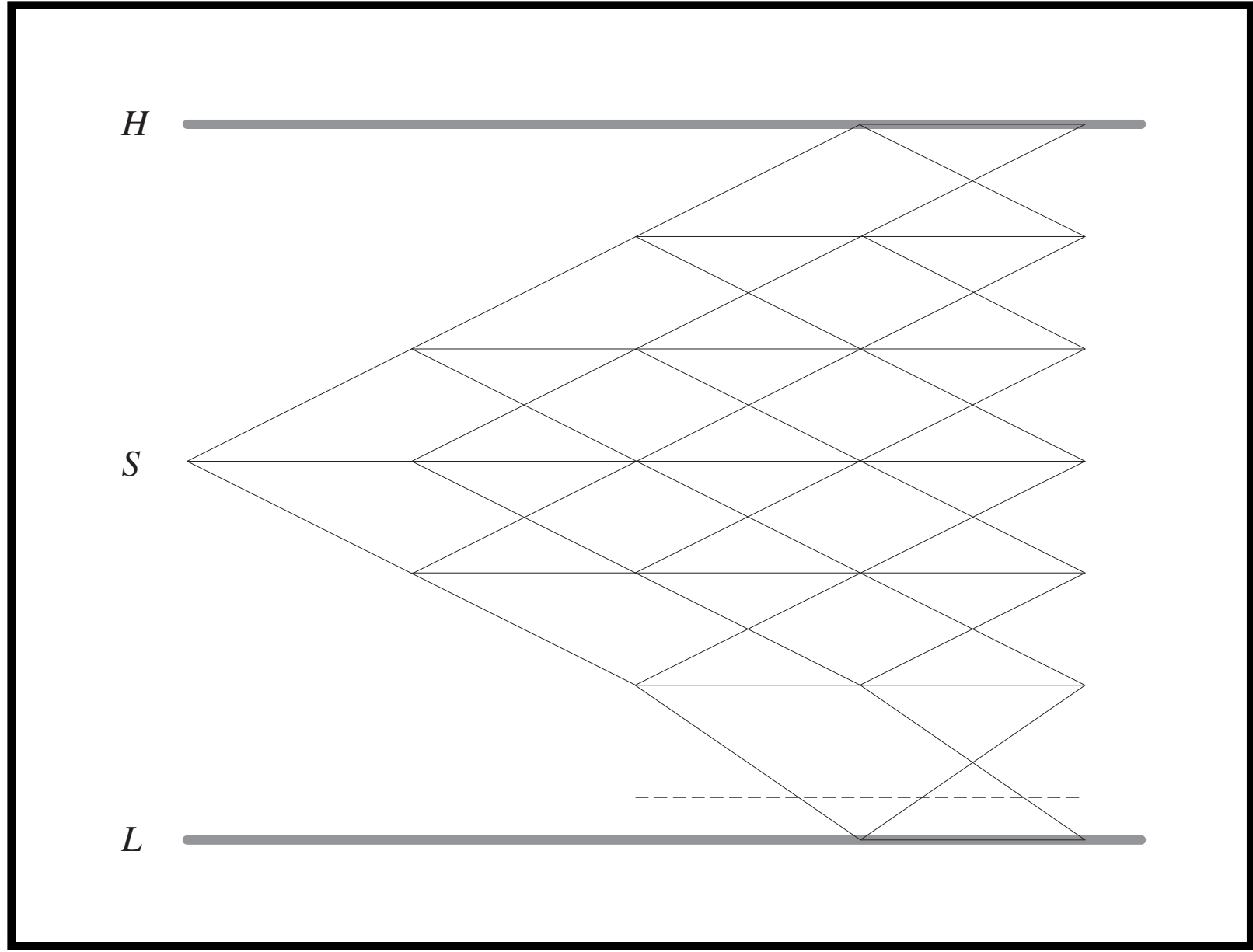
## Double-Barrier Knock-Out Options

- We knew how to pick the  $\lambda$  so that one of the layers of the trinomial tree coincides with one barrier, say  $H$ .
- This choice, however, does not guarantee that the other barrier,  $L$ , is also hit.
- One way to handle this problem is to lower the layer of the tree just above  $L$  to coincide with  $L$ .<sup>a</sup>
  - More general ways to make the trinomial model hit both barriers are available.<sup>b</sup>

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<sup>a</sup>Ritchken (1995).

<sup>b</sup>Hsu (R7526001, D89922012) and Lyuu (2006). Dai (R86526008, D8852600) and Lyuu (2006) combine binomial and trinomial trees to derive an  $O(n)$ -time algorithm for double-barrier options (see pp. 673ff).



## Double-Barrier Knock-Out Options (continued)

- The probabilities of the nodes on the layer above  $L$  must be adjusted.
- Let  $\ell$  be the positive integer such that

$$Sd^{\ell+1} < L < Sd^{\ell}.$$

- Hence the layer of the tree just above  $L$  has price  $Sd^{\ell}$ .<sup>a</sup>

---

<sup>a</sup>You probably cannot do the same thing for binomial models (why?).  
Thanks to a lively discussion on April 25, 2012.

## Double-Barrier Knock-Out Options (concluded)

- Define  $\gamma > 1$  as the number satisfying

$$L = Sd^{\ell-1}e^{-\gamma\lambda\sigma\sqrt{\Delta t}}.$$

- The prices between the barriers are

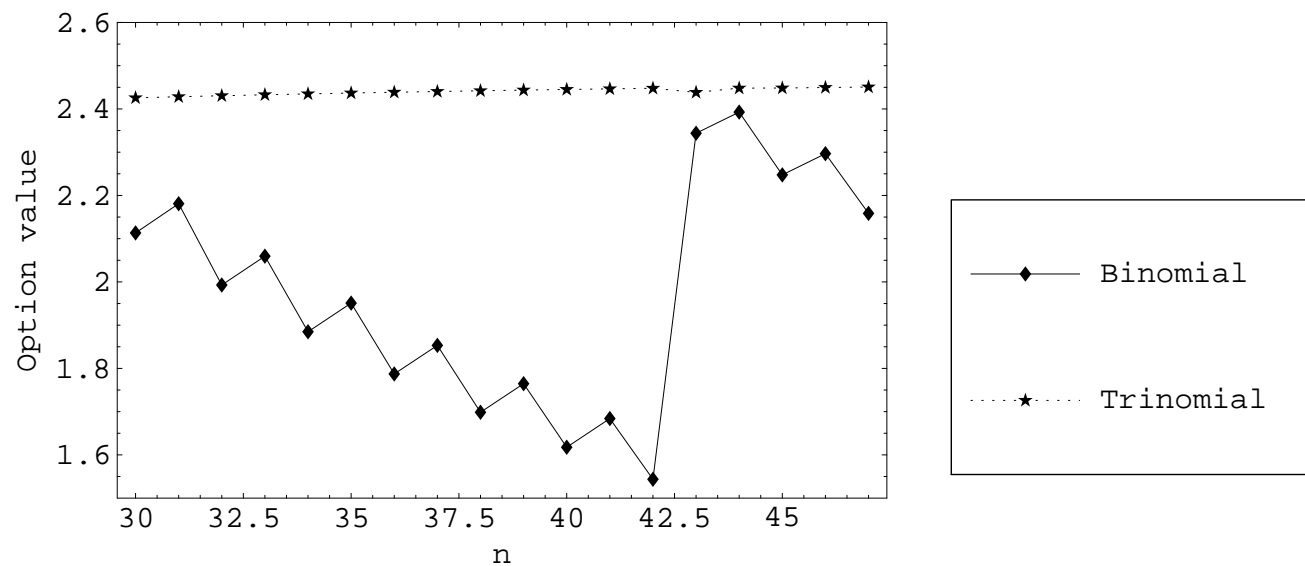
$$L, Sd^{\ell-1}, \dots, Sd^2, Sd, S, Su, Su^2, \dots, Su^{h-1}, Su^h = H.$$

- The probabilities for the nodes with price equal to  $Sd^{\ell-1}$  are

$$p'_u = \frac{b + a\gamma}{1 + \gamma}, \quad p'_d = \frac{b - a}{\gamma + \gamma^2}, \quad \text{and} \quad p'_m = 1 - p'_u - p'_d,$$

where  $a \equiv \mu'\sqrt{\Delta t}/(\lambda\sigma)$  and  $b \equiv 1/\lambda^2$ .

## Convergence: Binomial vs. Trinomial



## Ideas for Binomial Trees To Handle Two Barriers

